# Writing with Al

Al final project – team 26

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## Introduction

- Purpose
  - Generating text by trained model
- Computer is faster on heavy reading
- Inspiration
- Continue writing

#### Related work

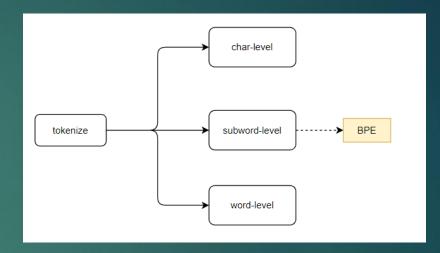
- ▶ BlinkDL/Al-Writer
  - ▶ RWKV Model
- Graykode/gpt-2-Pytorch
- huggingface/transformers
- mymuise/gpt2-quickly

#### Dataset

- ► Huyen Nguyen, SimpleBooks
  - ► Created from 1,573 Gutenberg books
  - ▶ 92M word-level tokens, on par with WikiText-103 (103M tokens)
  - ▶ vocabulary of 98K, a third of WikiText-103's.

#### Dataset - Construction

- ▶ BPE tokenizer
  - ▶ Why choose BPE tokenizer
  - Subword-level tokenizer
  - Keep mixing the highest probability characters into new subwords



```
u-n-<u>r-e</u>-l-a-t-e-d
u-n re-l-<u>a-t</u>-e-d
u-n re-l-at-<u>e-d</u>
<u>u-n</u> re-l-at-ed
un re-l-ated
un <u>re-l</u>-ated
un-related
unrelated
```

# Dataset – Construction (2)

- Normalizer
  - sentencepiece
  - ▶ NFKC
  - Improving operation efficiency
- Pre-tokenizer
  - Bytelevel
  - Replacing bytes of given string, Spliting into words.
- Special tokens

```
trainer = BpeTrainer(vocab_size=50000,
    show_progress=True,
    special_tokens=[
        "<s>",
        "<pad>>",
        "</s>",
```

# Dataset – Construction (3)

- Making Dataset
  - Load tokenizer
  - ► Encode the text
- Shuffle
  - Randomly shuffles elements of the dataset
- Batch
  - Combine elements of the dataset into batch
- Make Dataset
  - ► Type: tf.data.Dataset

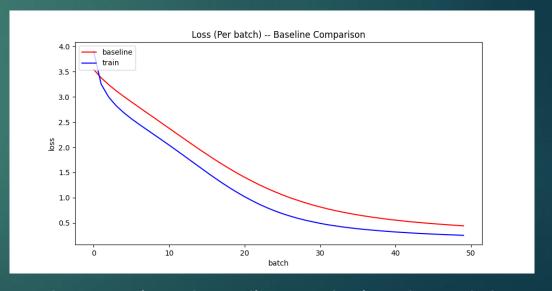
```
# Use tokenizer to encode the data as single string
data tokenized : str
data pos = config.train pos if type == "train" else config.test pos
with open(data pos, "r", encoding='utf-8') as f:
    tmp = f.read()
   tmp = tmp.replace("\n", " ") # clean up the change line characters
# Add token, this can be done when having multiple input files or text
# add special tokoen = tokenizer.bos token + tmp + tokenizer.eos token
data tokenized = tokenizer.encode(tmp)
print("\tText Encoded...")
# Slice data with equal quantity
examples, inputs, labels = [], [], []
for i in range(0, len(data tokenized) - config.block size + 1, config.block
    examples.append(data_tokenized[i:i + config.block_size])
for ex in examples:
    inputs.append(ex[:-1])
    labels.append(ex[1:])
dataset = tf.data.Dataset.from tensor slices((inputs, labels))
# Shuffles & Batch
dataset = dataset.shuffle(buffer size=config.buffer size) \
    .batch(config.batch size, drop remainder=True)
return dataset
```

train.py – making dataset

#### Baselines

- Comparison with other's program
  - mymuise/gpt2-quickly
- mymuise/gpt2-quickly
  - WordPiece Tokenizer
  - Without normalizer

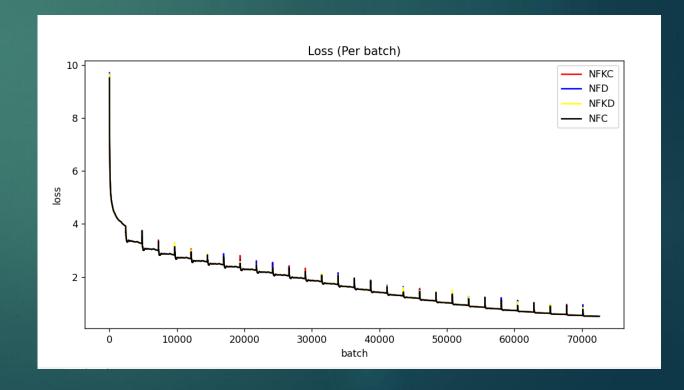
gpt2-quickly: build\_tokenizer.py



Comparison baselines & trained model

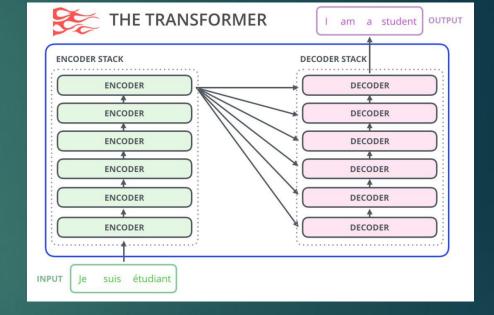
### Baselines

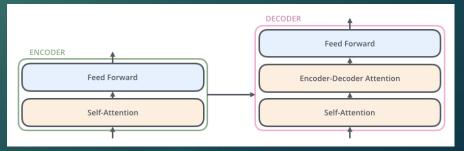
- Comparison with other normalizers
  - ▶ No noticeable difference
- ▶ Different Normalization Form
  - ▶ NFD
  - ▶ NFC
  - ▶ NFKC
  - ▶ NFKD



# Main Approach

- ▶ GPT-2
  - ► Generative Pre-trained Transformer
  - ▶ BPE Encoding
  - ▶ Transformer
  - ► Auto-regression
  - ▶ Self-Attention





# Main Approach – Code Structure

- ▶ src/\*
  - config.py: Program configuration
  - model.py: Model definition
  - ▶ tokenization.py: Build BPE tokenizer.
- train.py
  - Main procedure of training
- write.py
  - Generating text via pre-trained model
- Procedure
  - tokenization.py -> train.py

```
- src
- config.py
- model.py
- tokenization.py
- train.py
- write.py
```

Code structure

# Main Approach

- Byte-Pair Encoding
  - Using tokenizer built in former
  - translate text into numeric indices
- Construct the Model
  - Provided dataset
  - Callback function
- ▶ Text-generation
  - Top-k sampling
  - Redistributed on top K words

```
## Init Model ##
print("==> Init Model:")
train_model = TextModel(config=config, tokenizer=BPE_tokenizer, model_name="train")
test_model = TextModel(config=config, tokenizer=test_BPE_tokenizer, model_name="test")
print("\tModel initialized...")
```

#### Model initialization

Text generation with top-k sampling

#### Evaluation Metric

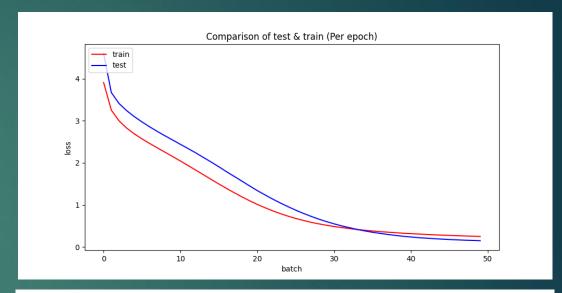
- ▶ Loss function
  - Sparse Categorical Cross Entropy
- Metric
  - Accuracy
- Optimizer
  - ▶ We choose Adam optimizer
  - Compared with
    - ▶ SGD
    - ▶ Momentum
    - AdaGrad

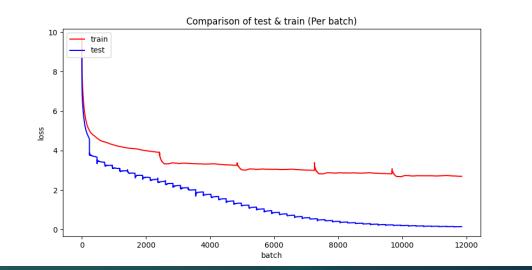
```
# Define
self.optimizer = tf.keras.optimizers.Adam(learning_rate=5e-5, epsilon=1e-08, clipnorm=1.0)
self.loss = tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True)
self.metric = tf.keras.metrics.SparseCategoricalAccuracy('accuracy')
```

Defined in model.py, Init of TextModel

# Evaluation Metric – Train/Test Split

- Separate train, test dataset
  - Validate model hyperparameters
  - ▶ Test Dataset
    - About 10% amount of Train Dataset
    - ▶ No cross data



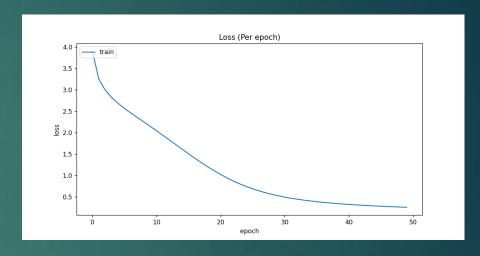


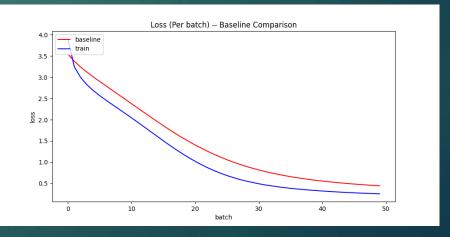
# Result & Analysis

- Baseline
  - Modified from mymuise/gpt2-quickly
- ▶ Text Generation
  - Cut on the max length
  - ▶ Stable character name

```
Training result output....
Input: Did you hear that
Setting `pad_token_id` to 2 (first `eos_token_id`) to generate sequence
Result:
```

Did you hear that "I believe the gypsies are trying to find ou t about this thing, "said Bert. "If they did they 'd have gone away. However, I 'll take you home with me for a little while, and then let my dog go away with us. "Perhaps the gypsies did, "said Mr. Bobbsey. "They took Helen's doll and were camping on the island of Lakeport. "I'll ask about that when they' ve camped on Blueberry Island, "said Mr. Bobbsey. "I 'Cause though they did n ot let the gypsies know





### Future Work

- ► Fixed / Optimized Problem
  - ▶ Preprocess
  - More dataset
  - ▶ Different Language
  - ► Collect general sentence ending

Thanks for your listening!